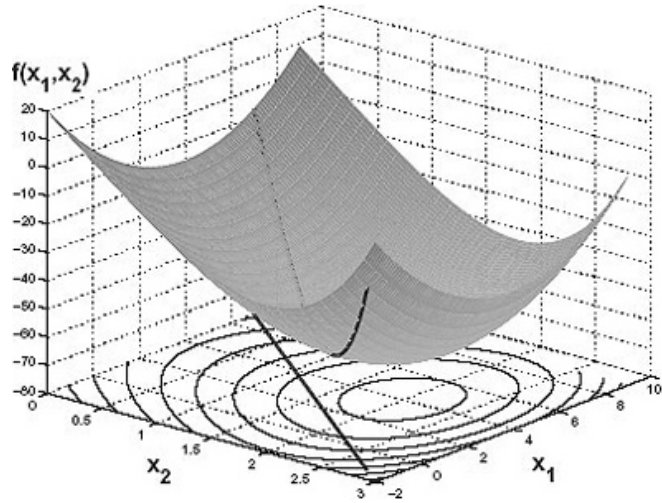


Multivariable problem

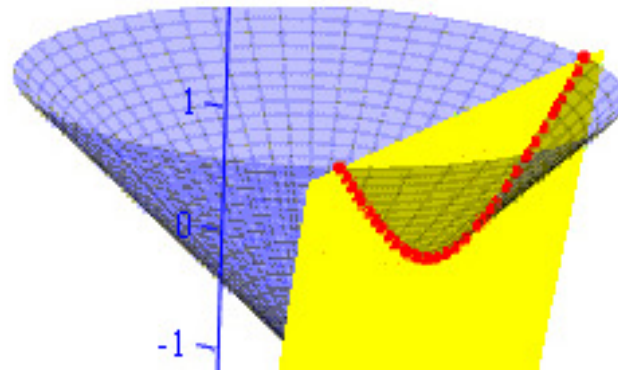
Rajib Kumar Bhattacharjya

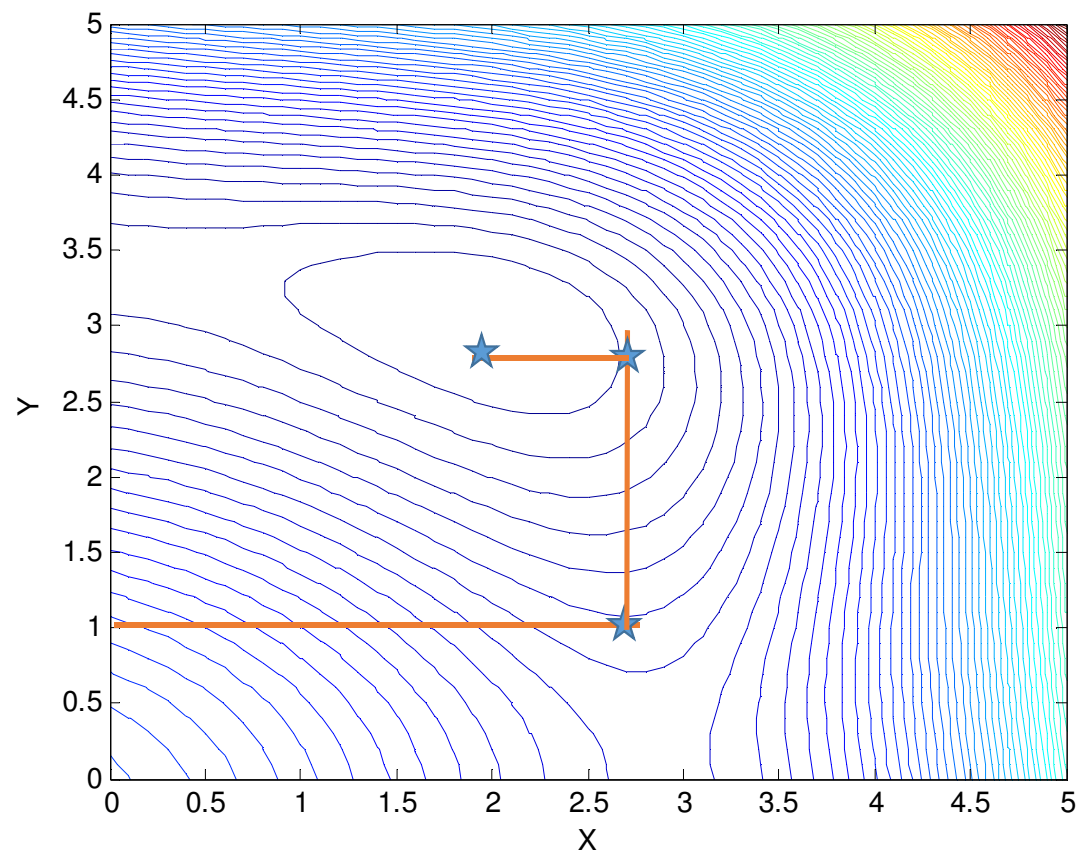
Department of Civil Engineering

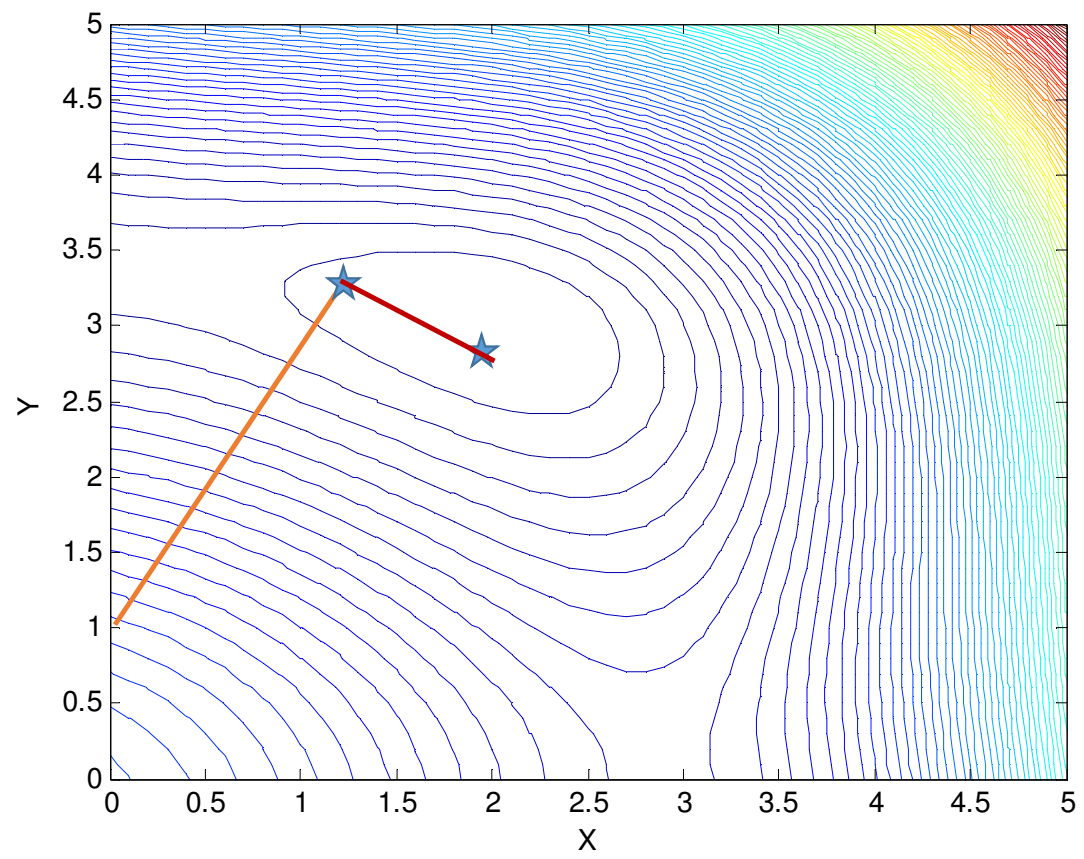
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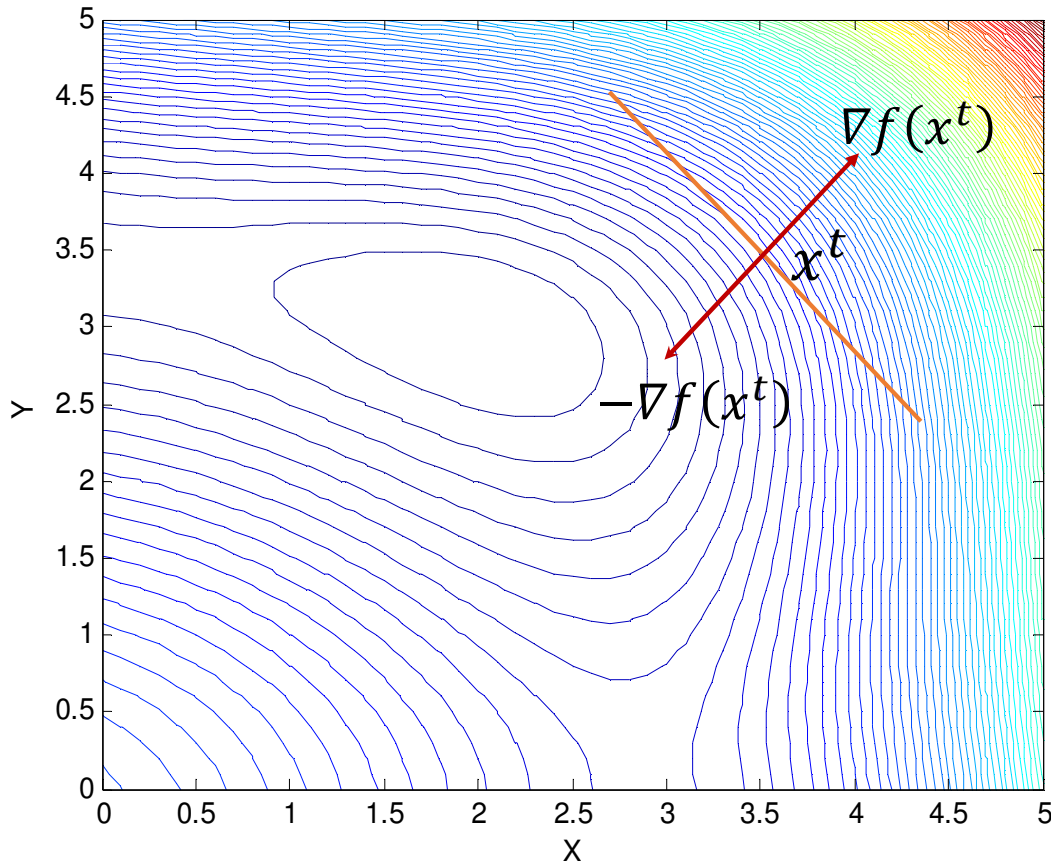


G/R. K. Chattachariya









Descent direction

A search direction d^t is a descent direction at point x^t if the condition $\nabla f(x^t) \cdot d^t < 0$ is satisfied in the vicinity of the point x^t .

$$\begin{aligned} f(x^{t+1}) &= f(x^t + \alpha d^t) \\ &= f(x^t) + \alpha \nabla^T f(x^t) \cdot d^t \end{aligned}$$

The $f(x^{t+1}) < f(x^t)$

When $\alpha \nabla^T f(x^t) \cdot d^t < 0$

Or, $\nabla^T f(x^t) \cdot d^t < 0$

Q. Show that the Newton's method finds the minimum of a quadratic function in one iteration

A quadratic function can be written as

$$f(X) = \frac{1}{2}X^TAX + B^TX + C$$

The minimum of the function is given by

$$\nabla f(X) = AX + B = 0$$

$$X = -A^{-1}B$$

Now apply Newton's method. The iterative step gives

$$X_{i+1} = X_i - H^{-1}\nabla f(X_i)$$

In this case $H = A$

$$X_{i+1} = X_i - A^{-1}(AX_i + B)$$

$$X_{i+1} = -A^{-1}B$$

$$\frac{\partial(X^TAX)}{\partial X} = AX + A^TX$$

In this case $A = A^T$

$$\frac{\partial(X^TAX)}{\partial X} = 2AX$$

$$\frac{\partial(AX)}{\partial X} = A^T$$

$$\frac{\partial(X^TA)}{\partial X} = A$$

$$\frac{\partial(A^TX)}{\partial X} = A$$

Powell's conjugate direction method

Parallel subspace property

Given a quadratic function $f(X) = \frac{1}{2}X^TAX + B^TX + C$ of two variables and X^1 and X^2 are the two arbitrary but distinct points.

If Y^1 is the solution of the problem

$$\text{Min } f(X^1 + \lambda d)$$

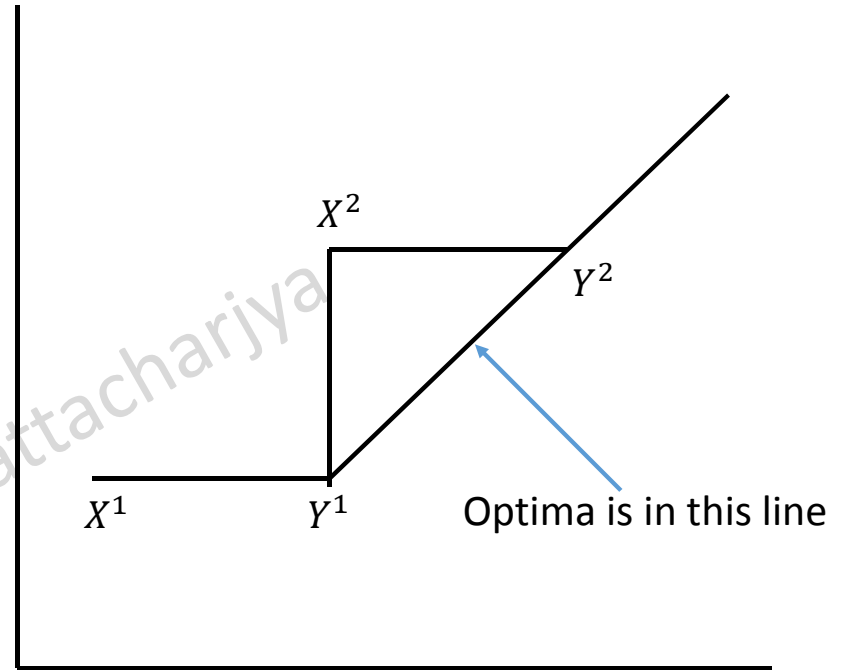
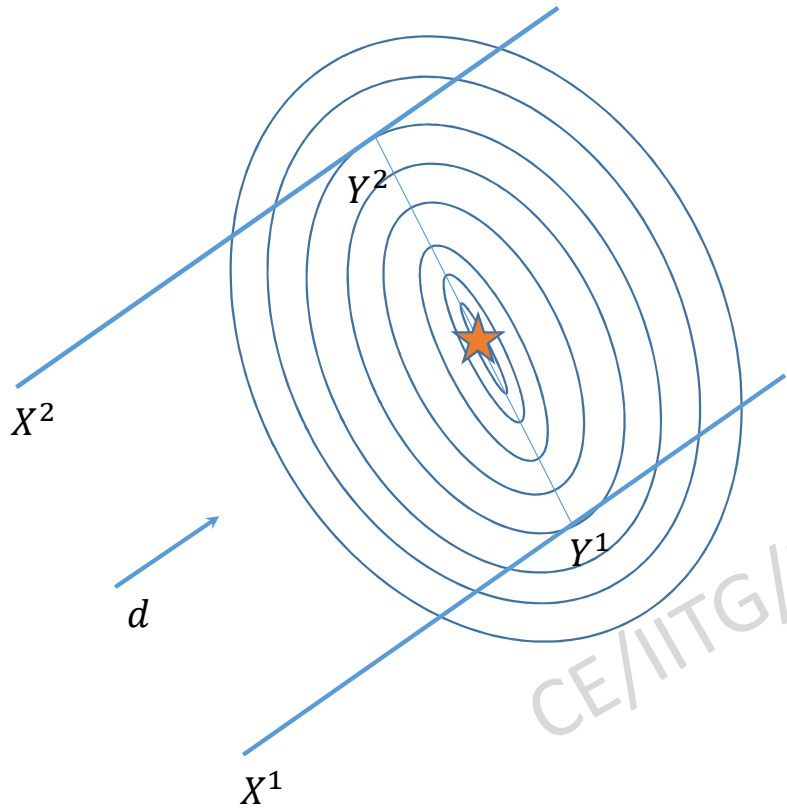
If Y^2 is the solution of the problem

$$\text{Min } f(X^2 + \lambda d)$$

Then the direction $(Y^2 - Y^1)$ is conjugate to d , or other words, the quantity

$$(Y^2 - Y^1)^T Ad = 0$$

For quadratic function minimum lies on the direction $(Y^2 - Y^1)$



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